

Residential Space Heating Determinants and Supply-side Restrictions: Discrete
Choice ApproachElena Stolyarova^{1,2,*}, Hélène Le Cadre², Dominique Osso¹, Benoit Allibe¹, Nadia Maïzi²

Abstract

This paper provides an empirical analysis of the supply-side constraints that impact household choices of space heating systems in French dwellings. Based on data from the 2006 National Housing Survey and the 2013 Household Survey, we estimate discrete choice models using socio-demographic, dwelling and spatial characteristics as determinants of choice. In order to capture the supply-side constraints, we perform a post-estimation clustering based on the Expectation-Maximization algorithm, which allows us to determine the groups of households that are most likely to choose each considered space heating system. The results suggest that for 2006, households living in individual houses preferred to heat space using an individual boiler, whereas those living in apartments opted for direct electric heating. In 2013 all households preferred direct electric heating, and wood heating became their second choice. The discrete choice models do not show a significant change in behavior from 2006 to 2013. The post-estimation clustering indicates that many households were strongly constrained by the conditions and characteristics of their living space in 2006. The mean probability of choosing direct electric heating or a boiler was more than 0.8 for 21% of households living in houses and for 36% living in apartments. Thus, these households did not seem to have a choice. In 2013 the supply-side constraints were weaker. The highest mean probability (0.755) is observed in the group opting for an individual boiler. We also observe an increase in the popularity of heat pump and wood heating systems.

Keywords: Dwelling Energy choice, Household Behavior, Multinomial Logit, Space Heating Systems

JEL: C25, C46, D12, D19, Q40, Q55

1. Introduction

French residential consumption represents a significant 26.5% (including wood) of total final energy consumption, which stood at 154.4 Mtoe³ in 2013 (ADEME, 2014a). The largest share of this consumption, between 50% and 70% on average, is used for space heating (ADEME, 2014a). The French government set a target to reduce energy consumption in the building sector by at least 38% from the 2006 baseline by 2020⁴. To achieve this target, two measures have been introduced into the residential sector: the new thermal regulation oversees energy efficiency retrofit measures and the government has created several types of financial aid to encourage households to undertake retrofit work. Despite these measures, energy consumption is decreasing too slowly: a drop of 2.6% in 2013 and only 1.4% from 2000 to 2010 (ADEME, 2014a; CEREN, 2011). French residential statistics show that dwellings built prior to the first thermal regulation of 1974 represent the most significant energy-saving potential. According to ADEME⁵, dwellings built before 1974 account for 55% of building stock and consume 64% of total residential energy consumption, while those built after 1998 account for 16% and represent only 11% of energy consumption. However,

*Corresponding author

Email address: elena.stolyarova@edf.fr (Elena Stolyarova)

¹EDF R&D, ENERGie dans les BATiments et les Territoires, av.des Renardières Ecuelles, 77818, Moret sur Loing, France

²MINES ParisTech, PSL Research University, CMA, CS 10207 rue Claude Daunesse, 06904, Sophia-Antipolis, France

³ Million tons of Oil Equivalent.

4SOeS

⁵French Environment and Energy Management Agency (in French: Agence de l'environnement et de la maîtrise de l'énergie).

part of this potential remains underexploited for two main reasons, i.e. slow increase of building stock, at only +1% new dwellings per year, and low level of retrofit works, which are insufficient to reduce energy consumption to meet an ambitious target of a 38% reduction in energy consumption in the building sector⁶.

To improve this situation, it is important not only to understand the main drivers of household behavior concerning energy consumption in the dwelling, but also to identify which supply-side barriers to overcome. For example, dwellings do not always have access to all energy sources. Natural gas can only be used in buildings connected to the grid, so that dwelling that is not located close to the gas pipeline requires major financial investment. Only 28% of French municipalities⁷ and 49.8% of dwellings had access to the gas grid in 2006 (INSEE, 2006). Wood heating is usually used in individual houses with free and ready access to wood. Moreover, fuel oil, wood and liquefied petroleum gas (LPG) need space for storage. Households cannot instantly switch from one source to another because Residential Space Heating Systems (RSHS) and domestic hot water generally use a single energy. A household that wants to change energy source must therefore buy and install new equipment. The dwelling occupation status, i.e. owner or tenant, also impacts possible retrofit actions.

The contribution of this study to the existing discrete choice framework on the determinants of residential energy behavior is threefold. First, we analyze space-heating determinants from the perspective of the socio-demographic, dwelling and spatial barriers that need to be overcome, since households cannot instantly change these characteristics. Second, we use post-estimation clustering to try to determine the household groups for which the degree of supply-side constraints is the highest. Le Cadre et al. (2009) propose to use the distribution of individual estimated probabilities from a discrete choice model in order to divide the population into groups. Following the same logic, we carry out a post-estimation clustering based on an Expectation-Maximization algorithm in order to determine which households prefer which heating system. Finally, we compare two years and analyze the change in supply-side constraints.

The rest of this paper is organized as follows. We offer a brief literature review on the discrete choice approach in Section 2. Section 3 describes the econometric approach and methodology for unsupervised clustering, followed by section 4 in which we present the model specification. The data and some descriptive statistics are presented in Section 5. In Section 6 we provide the estimation results for the choice of RSHS. Section 7 presents the results of post-estimation clustering. Finally, some concluding remarks are made in Section 8.

2. Literature review

In this study, we concentrate on micro econometric analysis, especially discrete choice models, which aim to display, predict and explain the choice between two or more mutually exclusive discrete alternatives. We can group recent works that use a discrete choice approach in the energy sector into 3 subjects.

The first type of study analyzes the impact of the consumer situation (socio-demographic, dwelling and spatial characteristics) on choices related to energy topics in dwellings. For example, Scasny and Urban (2009) investigate the impact of socio-demographic and dwelling characteristics on the presence or absence of various appliances in the home, such as dishwashers, fridges and microwaves in OECD countries. They also try to determine the number of appliances by type and the household energy-saving activities. Ameli and Brandt (2014) attempt to determine the social characteristics and behavior attitudes that explain the decision to invest in energy efficiency equipment (light, appliances, RSHS) and retrofit work. Ricci and Bérangère (2014) investigate which household conditions push French energy-vulnerable consumers into fuel poverty. Michelsen and Madlener (2012) analyze the influence of RSHS attributes and household situation on the decision to adopt innovative residential heating system. Braun (2010) proposes a space-heating choice model for Germany that focuses on dwelling types and regional differences. Rahut et al. (2014) examine the impact of socio-demographic variables on the choice of energy source in Bhutan.

The second type of article looks at the level of expected energy consumption, which is estimated simultaneously by discrete choice models and linear regression models. This framework, called discrete-continuous choice models, was proposed by Dubin and McFadden (1984), who at the same time investigate the demand for appliances and the demand for electricity by appliances. Using the same type of model, Nesbakken (1999), Vaage (2000) estimate fuel energy choices and consumption in Norway. Couture et al. (2012) look at wood-space heating choices and consumption in

⁶French Ministry of Ecology, Sustainable Development and Energy

⁷GDF Suez

the French Midi-Pyrenes region. Mansur et al. (2008) estimate an American energy model to determine the choice and consumption of fuel oil in buildings by both households and firms. They take into account the impact on climate. Liao and Chang (2002) investigate the energy demand for space heating and domestic hot water uses in the United States with a focus on old people.

Lastly, the third type of study use stated preferences data from a choice experiment survey. This data type is subject to a wide range of economic interpretations (e.g. willingness to pay, reservation prices, consumer surplus, etc.). Moreover, researchers can use this data to obtain information on alternatives not chosen by consumers and on alternatives that do not yet exist. Islam and Meade (2013) study the conditions for photovoltaic adaptation by Canadian households. Achtnicht (2011), Achtnicht and Madlener (2014) analyze how environmental benefits impact the choice of heating system and retrofit works in Germany. Banfi et al. (2008) estimate consumers willingness to pay for energy-saving measures in Swiss residential dwellings for both tenants and owners. Rouvinen and Matero (2013) perform stated choice experiment on private homeowners choice of heating system in Finland. The results show that Finnish households prefer ground-source heating and district heating, and that existing heating systems and dwelling characteristics have a significant impact on choices.

3. Methods

3.1. Discrete choice framework

The discrete choice approach, or Random Utility Model (RUM), was initially proposed by Thurstone (1927) and later adapted to economics by McFadden (1974). The RUM is composed of two parts. One is observable by the researcher and can be estimated, the other is considered to be random. The RUM allows us to obtain the probability by choice alternative and by household. The utility function for choice j among J mutually exclusive choice alternatives on N households can be described as:

$$U_{jn} = V_{jn} + \xi_{jn}, \forall n = 1, \dots, N, \forall j = 1, \dots, J \quad (1)$$

Where U_{jn} is the utility of choice j for household n , V_{jn} is the determinist part of utility observable by the researcher, and ξ_{jn} is a stochastic random variable comprising unobservable household and dwelling characteristics, as well as measurement errors. To estimate the determinist part of utility, the researchers suppose V_{jn} to be linear in parameters:

$$V_{jn} = \beta^T X_{jn} \quad (2)$$

The vector of unknown parameters to estimate is expressed in β where T stands for the transpose operator. X_{jn} represents the matrix of explanatory variables. In this study we decided to use a standard Multinomial Logit Model (MNL) where ξ_{jn} is independent and identically distributed following the Gumbel Extreme Value distribution. The density function $f(\cdot)$ and the cumulative distribution function $F(\cdot)$ for each unobserved component of utility:

$$f(\xi_{jn}) = e^{-\frac{\pi^2}{6}} e^{-e^{\xi_{jn}}} \quad (3)$$

$$F(\xi_{jn}) = e^{-e^{\xi_{jn}}} \quad (4)$$

The difference between two Gumbel variables gives a logistic distribution:

$$F(\xi_{jn} - \xi_{in}) = \frac{e^{\xi_{jn} - \xi_{in}}}{1 + e^{\xi_{jn} - \xi_{in}}} \quad (5)$$

The probability that household n chooses choice alternative j is inferred from equation 5 given by a closed form expression in equation 6 and fitted by Maximum Likelihood estimation.

$$P(\xi_{jn}) = \frac{e^{\beta^T X_{jn}}}{\sum_{i=1}^J e^{\beta^T X_{in}}} \quad (6)$$

The estimated parameters give us the impact of the explanatory variables on the probability of choosing one choice alternative compared to the reference (or base) alternative and to the reference level of categorical variable. The use of MNL was also spurred by the nature of the dataset, which only allows us to estimate alternative specific individual parameters. Note that both datasets are revealed preference data, i.e. we know the final choice or situation of households but we have no information about the characteristics of the alternatives not chosen.

3.2. Methodology for unsupervised clustering

Assuming that the estimated probability distribution is a Gaussian mixture, we can use the Expectation-Maximization (EM) algorithm to cluster the households into groups. The EM algorithm was proposed by Dempster et al. (1977). This is a two-step algorithm. In the first step, we compute the expected value of the log likelihood function (normal distribution function in our case). In the second step, maximization, we find the parameters of distribution that maximize the log likelihood quantity. To perform the clustering we use a *mclust* package for R (Fraley et al., 2012). For univariate data, as in our case, the package proposes two types of Gaussian mixture, i.e. with equal variance for all normal distribution in the mixture, and different variance for each distribution. The number of classes for clustering and the type of mixture are chosen in order to minimize the Bayesian Information Criterion (BIC) (Schwarz, 1978).

4. Model specification

The concept of household choice of RSHS is complicated and depends on the households situation. A household can choose the type of RSHS and energy source at the time of construction if it is buying a new dwelling. In this case, the household is limited only by its income, preferences and access to energy sources. Alternatively, a household can choose the RSHS indirectly at the time of purchase (owner) or when signing the rental contract (tenant). However, in the case of social tenants (HLM⁸ or other), households do not choose their dwelling, which in France is allocated by a commission for each HLM body after analyzing households situations. Once an RSHS has been installed, it is difficult to change its type. Boilers, district heating, air/ water and water/water heat pumps require water loop networks. Direct electric heating requires radiative, convection or other types of heater. Wood heating systems (chimneys, wood burning stoves, etc.) need space to store wood and a network for hot air distribution in all dwelling. These network costs add to the cost of RSHS installation.

We must also consider who takes the decision to perform retrofit work in dwellings. In individual houses, owners decide to initiate retrofit work, either for themselves or their tenants. In the case of co-owned dwellings, all owners need to make the decision. In France, the average duration, from the point when co-owners decide to undertake space heating work to the point when the work is completed, is estimated at 4 to 5 years (ADEME, 2014b). Finally, we must point out the importance of household preferences at the time of choosing an RSHS, such as: energy type, type of heater, product brand, cost, government grant, information about RSHS type, etc.

The age of dwelling seems to be an important determinant in household choices. The French government has raised thermal standards for buildings since 1974⁹. The level of insulation also impacts on choice because energy bills are higher in dwelling with no insulation than in insulated dwellings. Household size is correlated to dwelling size, and thus increases annual space-heating bills. In order to separate the effect of dwelling size from household size, we use both variables as determinants of choice. The absence of access to the gas grid limits household choice for fuel used in boilers. The use of fuel oil and LPG in boilers is costly, and electric and wood boilers are not widespread. Therefore, if households do not have access to the gas grid, the likelihood of installing a boiler decreases.

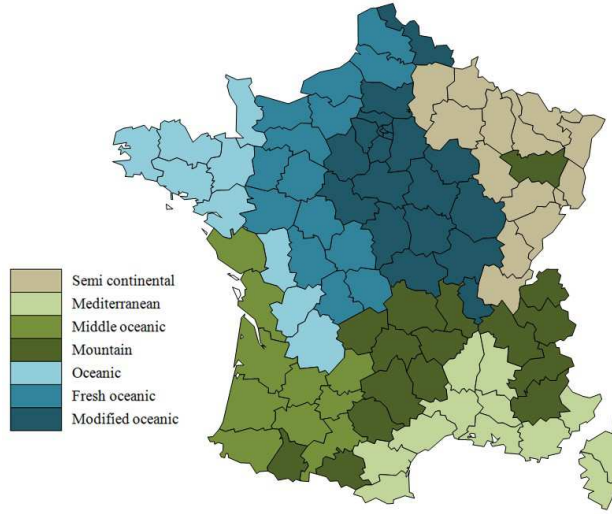
To take into account the climate variability in France, we decided to use the climate zone as an explanatory variable. Another possible option is to use the normalized heating degree days by French *département*. However, this measure focuses exclusively on the outdoor temperature. Other climate characteristics can impact the choice of RSHS, such as: the level of precipitation, the cycles and rhythms, number and duration of extreme events per year (storm, snow, heatwave, etc.). Households may also choose the location of their residence in line with individual climate preferences. We integrate into our model 7 climate zones as defined in the National Housing Survey (2006): oceanic, fresh oceanic, middle oceanic, modified oceanic, semi-continental, Mediterranean and mountain (Figure 1). As the climate zone does not consider the landscape of the residential area, we introduce urban density as a determinant of choice. Dwellings in rural zones have easier access to wood, but more difficulties accessing the gas grid.

Energy prices and the cost of space-heating installation also affect household choice. The data type used in our study does not give us information about the age of RSHS and we cannot propose an approximate cost of installation due to high variation of installation costs (Laurent et al., 2011). Moreover the mean costs for each type of RSHS are

⁸Rent-controlled housing (in French: *Habitation à Loyer Modéré*)

⁹The first thermal regulation was introduced in 1974, revised in 1988, 2000, 2005 and in 2012.

Figure 1: Climate zone location as defined in NHS 2006 (INSEE, 2006)



not the same from one database to another (Laurent et al., 2007). To input energy prices into the models, we assume that the heating system is chosen when the household moves into a new dwelling.

Finally, socio-demographic factors can impact household choice. We use only income level in our estimations. We tried to implement other social and economic characteristics: social professional category of household reference person and his/her partner, highest level of education, main occupation (employment status, student, pensioner, etc.). However, these are not significant for two reasons: first, these characteristics tend to evolve from year to year, whereas the energy choice covers a longer period; second, these characteristics are correlated with the occupancy status, income level, size of dwelling and the age of the household reference person.

In our study we use Revealed Preferences data, which only gives us information about the RSHS chosen and a description of the household situation. We propose to consider the households' situation at the time of the surveys as a result of consecutive choices. These results can be summarized by the final situation: household dwellings possess accurate RSHS. Households are assumed to have chosen the situation that maximizes their utility with respect to their income, the characteristics of the dwelling and energy prices. Thus, we can investigate the impact of the household environment on their dwelling space heating profile. We consider the most common, widespread heating system in France: space heating by Direct Electric Heating (DEH) and boiler (individual or collective). We also add the choice of a heat pump and wood heating system (chimney, wood burning stove) to take into account systems based on renewable energy.

5. Data

To analyze the year 2006, we used the most recent National Housing Survey (NHS 2006) published by INSEE¹⁰ which describes French housing expenditure and household living conditions (INSEE, 2006). The NHS 2006 is a large survey¹¹ covering 42,701 main residences. The survey provides detailed data on building characteristics, household living conditions (legal measures for occupation, difficulties in access to housing, household residential mobility, etc.), energy expenditure and consumption of wood. We can clearly identify the energy source for different uses and the type of heating system. We were forced to reduce the dataset from 42,701 to 32,305 dwellings due to lack of information on energy expenditure, consumption and RSHS for French overseas territories. We also removed dwellings with minority heating systems that are not part of our study. We took out dwellings that were neither houses nor apartments, and those for which the housing occupancy status was neither owner nor tenant.

¹⁰French National Institute for Statistics and Economic Research (in French: *Institut National de la Statistique et des Études Économiques*)

¹¹The next NHS is scheduled for 2016

The second dataset is a 2013 Housing Survey (HS 2013) conducted by TNS SOFRES on behalf of EDF¹² and covering 1,350 main French residences after data clearing. The HS 2013 data is a detailed survey covering energy expenditure, energy equipment for home end-uses, appliances, lighting, energy consumption, electricity tariffs, dwelling and household characteristics.

We also used an additional dataset of energy prices (in euro at current prices) by dwelling characteristic and by year from 1983 to 2006. Energy prices were taken from a French database Pégase (Ministère de l'écologie, du développement durable et de l'énergie, 2014). As we are examining the choice of RSHS, we take the tariff corresponding to high energy consumption. We choose 3 *in and off peak rates* for electricity: the price for small apartments¹³ with 6 kVA power, the price for other apartments with 9 kVA power and the price for individual houses with 12 kVA power. For natural gas, we consider the rate for apartments and the rate for houses. For fuel oil we do not have much choice, because the Pégase database only gives the rate for individual houses. The price is given in euro by 100 kWh for electricity, and by 100 kWh LCV¹⁴ for natural gas and fuel oil.

The descriptive statistics of choice determinants are given in Table 1 and Table 2. The map of climate zones is displayed in Figure 1. Access to the gas grid and roof insulation are not comparable between the two datasets. HNS 2006 gives us information on access to the gas grid for each dwelling, but HS 2013 only provides information on towns access to the gas grid. There are 36,614 municipal areas in France, 72% of which do not have access to gas grid. We note that in HS 2013, only 18.2% of municipal areas have no access to the gas grid. This percentage is explained by the design of the survey.

¹²Electricité de France

¹³Small apartment - apartment with living area under 35 square meters

¹⁴Lower Calorific Value

Table 1: Summary of the socio-demographic and dwelling characteristics

	Year	
	2006	2013
No access to gas grid	50.2 ¹	18.2 ²
No double-glazing	29	14.7
No roof insulation	89.2 ³	49.7
Income per month in euro		
< 1200	22.3	13.5
1201–1900	21.2	27.9
1901–2700	19.7	21.6
2701–3800	18.9	22
> 3800	17.9	15
Occupancy status		
Free	3.5	1.5
Owner	60.8	58.5
Social housing tenant	15.8	17
Other tenant	19.9	23
Dwelling age		
pre-1974	55.1	47.9
1974–1989	25.6	25.3
1990–1998	10.6	9
post-1998	8.7	17.8
Dwelling type		
Individual house	56	57.2
Apartment	44	42.8
Urban density		
Rural	23.4	22.6
Urban	57.7	59.5
Paris area	18.9	17.9
Climate zone		
Mediterranean	12.5	14
Middle oceanic	10.9	9.5
Oceanic	9.0	10.7
Fresh oceanic	12.9	12.4
Modified oceanic	32.6	31
Semi-continental	13.1	10.8
Mountain	9.0	11.5
Choice of heating system		
Individual boiler	48	45
Collective boiler	20.1	11
Heat pump	1.3	7
Direct electric heating	28.9	29.5
Chimney and wood stove	1.7	7.5

¹ The dwelling has no access to the gas grid, but the town may have access.

² The town of residence has no access to the gas grid.

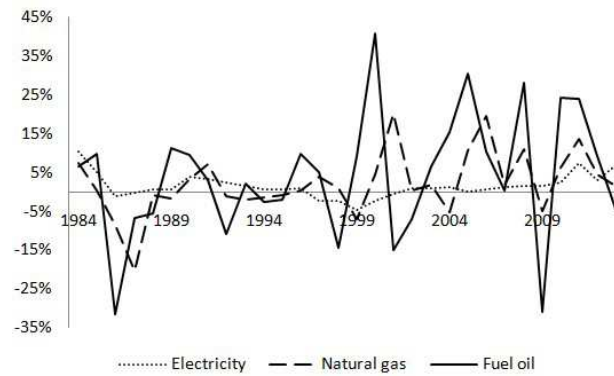
³ Only for individual houses.

Table 2: Summary of continuous variables

Variables	Min	Mean	Max	Std. dev.
Electricity price for small flat (Euros/100 kWh)	10.59	15.29	12.51	0.92
Electricity price for flat (euro/100 kWh)	9.84	14.34	11.95	0.83
Electricity price for house (euro/100 kWh)	9.13	14.06	11.31	0.93
Natural gas price for flat (euro/100 kWh LCV)	3.31	7.37	4.42	1.20
Natural gas price for house (euro/100 kWh LCV)	3.06	7.04	4.20	1.18
Fuel oil price for house (euro/100 kWh LCV)	2.80	9.72	4.72	2.03
Number of inhabitants 2006	1	2.3	14	1.27
Number of inhabitants 2013	1	2.16	6	1.11
Living area (m ²) 2006	1 ¹	93.5	999	46
Living area (m ²) 2013	15	97.5	350	41
Age of reference person 2006	16	52.3	101	17.4
Age of reference person 2013	20	52.2	97	16

¹ The living area is calculated only for dwelling areas above 1.80 meters high.

Figure 2: Growth rate of domestic energy prices from 1983 to 2013, Pégase 2014



For the remaining variables, we observe no significant differences between 2006 and 2013 for most dwelling and household characteristics, with three exceptions.

First, the penetration rate of double-glazing increases: 85.3% of dwellings have double-glazing in HS 2013 against 71% in NHS 2006. Next, in HS 2013 the number of dwellings built after 1998 is double that of NHS 2006. This is logical, since by definition, the period "After 1998" totals 15 years in HS 2013 against 8 years in NHS 2006. Finally, 27.9% of households have a monthly income of between 1,201 and 1,900 euro in 2013, compared to 21.2% in 2006. The share of low-income households (income lower than €1,200) drops from 22.3% to 13.5%. This change can be explained by an increase in the guaranteed monthly minimum net wage in France (S.M.I.C.¹⁵) from €984.61 in 2006 to €1121.71 in 2013.

More than half of households own their residence, 16% – 17% are social tenants (HLM), about 21% are non-social tenants and about 2% occupy dwelling free of charge. 57% of dwellings are individual houses, the rest are collective (apartments). Dwellings in urban areas account for 77.3%, one quarter of which is located in the city of Paris. One third of households live in modified oceanic climate zones. The average household comprises 2.2 people, with a living area of about 94 square meters. The average age of the reference household person is 52.2. Figure 2 shows the growth rates of domestic energy prices from 1983 to 2013. The price of fuel oil is the most volatile followed by natural gas. The growth rate of electricity prices is more stable, but it follows a similar trend to gas and fuel oil prices.

Regarding the frequencies of choice alternatives for 2006 and 2013, we can see that the number of households using heat pumps or wood as their main energy source for heating increased sharply in 2013 showing the increase of renewable energy use, which reduced the market share of boilers. However, the market share of DEH did not change between 2006 and 2013.

¹⁵Guaranteed minimum wage (in French: *Salaire Minimum Interprofessionnel de Croissance*)

6. Estimation results and discussion

We propose three models to quantify the choice of RSHS: in individual houses (2006), in apartments (2006) and for all dwellings in 2013¹⁶. All models have the following choice alternatives: DEH (base alternative), heat pump, or individual boiler¹⁷. In NHS 2006, we only use collective boilers as a choice in the apartments model, and only use wood heating systems in the individual houses model. However, HS 2013 shows a significant share of collective boilers used in houses and wood-burning stoves in apartments, which cannot be ignored. We therefore use both RSHSs in the same model.

To input the energy prices into the models, we assume that the RSHS is chosen when the household moves into a new dwelling. Only NHS 2006 provides us with information on the year that households moved into their homes. For each household, we take the electricity, natural gas and fuel oil prices for the year they moved in from a database Pégas 2014. Prices are available dating from 1983. This means that the parameters associated with energy prices only impact the choice probabilities for dwelling occupied by the same household since 1983. We try to bridge the lack of information by implementing the mean annual prices for the years before 1983, the growth rate of energy prices, and the same energy prices for dwellings without prices, but none of these configurations works. Moreover if we remove energy prices for all dwellings, the quality of the model decreases.

6.1. Results for both 2006 and 2013 years

The estimation results are given in Tables 3, 4 and 5. The estimation makes a satisfactory adjustment of data with pseudo \bar{r}^2 about 0.33 for 2006 models and 0.2 for 2013¹⁸. The estimation results clearly indicate that the preferred heating system depends on the dwelling type in 2006. Occupants of individual houses prefer to heat space using boilers, while occupants of apartments prefer DEH *ceteris paribus*. However, the least preferred heating system is the same the heat pump. If we look at the results for 2013, we note that the order of preference has changed. DEH is the preferred heating system for all dwelling types, followed by chimneys and wood burning stoves and ending with collective boilers. Households seem to have no preference between individual boilers and heat pumps when the value of constants is closed: -2.5 for individual boilers and -2.53 for heat pumps.

In 2006, the impact of fuel price is not significant in both models. The level of income, household size and age are not significant in the apartment model 2006. In 2013, income, dwelling occupation status, access to the gas grid and insulation are not significant. For other variables, we observe that an increase in household age and household size increases the probability of using a boiler, heat pump or wood heating in dwellings for both years. However, this impact is limited and the Relative Risk Ratio¹⁹ (RRR) never exceeds 2 with the highest RRR in 2013: the probability of using a collective boiler or heat pump is multiplied by 1.3-1.4 for each additional inhabitant. We also observe a similarity for climate variables in the model. DEH, heat pumps and wood heating are more prevalent in the Mediterranean climate zone. Wood heating is also a more likely choice in the semi-continental climate.

6.2. Individual houses in 2006

For 2006, the results have two important implications. The absence of access to the gas grid has the highest effect on choice probability in favor of DEH systems. The RRR is 0.03*** for individual boilers in individual houses and 0.014*** for individual boilers (0.07*** for collective boilers) in apartments. This means that if a dwelling has no access to the gas grid, the probability of having a boiler in a dwelling in 2006 was divided by 14–71. The lack of double-glazing multiplies the probability of choosing an individual boiler by 1.91 and of choosing wood heating by 1.637. The absence of roof insulation increases the probability of choosing wood with RRR 1.981***

The energy prices parameters have the expected sign. An increase in electricity prices encourages the use of boilers and wood heating systems and discourages the use of DEH and heat pumps for which electricity is an energy source. An increase in natural gas prices or fuel oil prices has a negative impact on the probability of choosing boilers and wood heating systems respectively. However, the RRR is close to 1 in individual houses for all choice alternatives.

¹⁶The size of sample is not sufficient in 2013 to estimate separate models by dwelling type

¹⁷Boiler specific to each dwelling

¹⁸The 0.2 of pseudo \bar{r}^2 is equivalent to 0.5-0.6 R^2 for linear regression models (McFadden and Domencich, 1975, p. 124)

¹⁹A relative risk ratio (RRR) of 2 means that one type of RSHS would be twice as likely to be chosen as another type of RSHS.

A 1 €/100 kWh increase in energy prices multiplies (or divides) the probability of using each RSHS by 1-1.13. High-income households prefer boilers and heat pumps, while low-income households prefer wood. More tenants use DEH. Boilers and wood are present in houses built prior to the first thermal regulation, whereas heat pumps are present in new dwellings.

6.3. Apartments in 2006

We observe several differences in the results of heating system choices in apartments (2006). First at all, gas prices, fuel oil prices, size of household, age of referent person and income do not seem to impact on choice probabilities. The coefficient for electricity prices is significant and negative, but has a limited impact on probability, which is divided by 1.23 (RRR 0.811 – 0.962). The size of dwelling increases the probability of choosing a boiler or heat pump, while the presence of double-glazing has a negative impact and encourages the use of DEH. The absence of double-glazing multiplies the probability of using a heat pump by 3.928. Boilers are more prevalent in semi-continental and mountain climate zones. If an apartment is located in a semi-continental climate, the probability of using a boiler is multiplied by 4.239 (individual) or by 6.374 (collective) *ceteris paribus*. Boilers are mainly present in HLMs with an RRR of between 2.049 (individual) and 5.456 (collective). Collective boilers are used in apartments built prior to 1974: the probability of having a collective boiler is divided by 5 in new buildings. Individual boilers are more prevalent in apartments built between 1974 and 1998. Similarly to the model for individual houses, heat pumps are prevalent in new dwellings. DEH is more widespread than other RSHSs in rural zones, with the exception of Parisian dwellings, where the probability of using an individual boiler is lower than in rural zones and other urban zones.

6.4. Dwellings in 2013

The model for 2013, which includes houses and apartments, reveals some differences. The size of dwelling and household, as well as the age of reference person, are significant and have a positive impact on the probability of choosing a boiler, heat pump or wood with an RRR close to 1. However, double-glazing, roof insulation, occupancy status and income are not significant. As in the 2006 model, heat pumps are most prevalent in the Mediterranean climate zone, while boilers and wood are more common in semi-continental or mountain climate zones. For dwelling built before first thermal regulation (1974), the probability of having an individual boiler, collective boiler or wood heating is divided on average by 2, 10 and 3 respectively. Heat pumps are more prevalent in new dwellings at RRR 1.465. The "apartment" dwelling type is more likely to have a collective boiler with RRR 51.8.

Table 3: Results of MNL for choice of heating system in individual house (2006)

Choice alternatives (Frequency of alternatives)	Individual boiler (64%)		Heat pump (1.5%)		Wood (3%)	
	Coefficient	RRR [#]	Coefficient	RRR [#]	Coefficient	RRR [#]
Constant	2.21***	NA	-5.817***	NA	-0.322***	NA
Electricity price	0.0387**	1.039**	-0.128***	0.879***	0.07**	1.072**
Gas price	-0.279***	0.756***	0.243	1.275	-0.046	0.954
Fuel price	0.0136	1.013	0.122	1.130	-0.22	0.801
Size of dwelling	0.012***	1.011***	0.015***	1.015***	0.0013	1.001
Household size	0.097***	1.101***	0.123**	1.131**	0.269***	1.309***
Age of RP	0.004**	1.004**	0.009	1.009	-0.011***	0.988***
No access to gas grid	-3.543***	0.029***	-0.505**	0.603**	-0.3813*	0.683*
No double-glazing	0.647***	1.910**	-0.116	0.889	0.493***	1.637***
No roof insulation	-0.007	0.992	-0.411	0.662	0.683***	1.981***
Climate zone						
Mediterranean	base	base	base	base	base	base
middle oceanic	0.114	1.122	-0.118	0.888	-0.23	0.794
oceanic	0.442***	1.557	-0.572***	0.564***	-0.664***	0.514***
fresh oceanic	0.605***	1.831	-0.498**	0.607**	-0.045	0.955
modified oceanic	0.378***	1.459	-0.779***	0.458***	-0.334*	0.715*
semi-continental	1.302***	3.679	-0.113	0.893	0.481***	1.618***
mountain	0.761***	2.139	-0.29	0.741	0.024	1.024
Income per month (euro)						
< 1200	base	base	base	base	base	base
1201–1900	0.189**	1.208	0.754*	2.125*	-0.122	0.884
1901–2700	0.198**	1.219	0.544	1.723	-0.388***	0.678***
2701–3800	0.134*	1.143	1.092***	2.987***	-0.83***	0.436***
> 3800	0.01	1.010	0.921**	2.512**	-1.737***	0.176***
Dwelling occupancy status						
owner	base	base	base	base	base	base
tenant of HLM	0.098	1.103	-15.01	0	-1.186***	0.305***
other tenant	-0.686***	0.503	-1.891***	0.150***	-0.942***	0.389***
free	-0.310**	0.733	-0.514	0.597	-0.188	0.828
Dwelling age						
before 1974	base	base	base	base	base	base
1974–1989	-1.058***	0.346	0.812***	2.254***	-1.304***	0.271***
1990–1998	-1.327***	0.265	-0.256	0.773	-0.893***	0.409***
after 1998	-1.124**	0.325	0.183	1.201	-0.924***	0.396***
Urban density						
rural	base	base	base	base	base	base
urban	-0.159***	0.852	0.14	1.150	-0.642***	0.525***
Paris area	-0.438***	0.645	-0.72*	0.483*	-1.815***	0.162***
Pseudo $\bar{\rho}^2$	0.3232					
Log-likelihood	-9789					
Observations	17618					
Base category: Direct Electric Heating						

Relative Risk Ratio of one preference over other

* Coefficient significant at 90 confidence level

** Coefficient significant at 95 confidence level

*** Coefficient significant at 99 confidence level

Table 4: Results of MNL for choice of heating system in apartments (2006)

Choice alternatives (Frequency of alternatives)	Individual boiler (30%)		Collective boiler (44%)		Heat pump (0.2%)	
	Coefficient	RRR [#]	Coefficient	RRR [#]	Coefficient	RRR [#]
Constant	-0.057***	NA	-0.879***	NA	-7.79***	NA
Electricity price	-0.038*	0.962*	-0.048**	0.952**	-0.209	0.811
Gas price	0.46	1.047	0.059	1.060	0.853	2.347
Fuel price	0.008	1.008	0.027	1.027	-0.151	0.859
Size of dwelling	0.02***	1.021***	0.016***	1.016***	0.019**	1.019**
Household size	-0.01	0.991	0.037	1.038	0.129	1.137
Age of RP	-0.01	0.996	0.003	1.003	0.002	1.002
No access to gas grid	-4.26***	0.014***	-2.637***	0.071***	-0.404	0.667
No double-glazing	0.416***	1.516***	0.906***	2.476***	1.368***	3.928***
Climate zone						
Mediterranean	base	base	base	base	base	base
middle oceanic	0.642***	1.901***	0.329**	1.390***	-0.685	0.503
oceanic	0.907***	2.478***	0.765***	2.150***	-13.84	0
fresh oceanic	1.09***	2.995***	1.612***	5.014***	-13.66	0
modified oceanic	0.764***	2.148***	1.25***	3.491***	-0.676	0.508
semi-continental	1.444***	4.239***	1.852***	6.374***	-13.29	0
mountain	1.355***	3.877***	1.861***	6.433***	0.64	0
Income per month (euro)						
< 1200	base	base	base	base	base	base
1201–1900	0.023	1.023	0.099	1.104	0.658	1.932
1901–2700	0.102	1.107	0.248***	1.282***	0.932	2.539
2701–3800	0.056	1.058	0.19*	1.210*	-0.06	0.939
> 3800	-0.006	0.993	0.022	1.022	0.4	1.493
Dwelling occupancy status						
owner	base	base	base	base	base	base
tenant of HLM	0.717***	2.049***	1.696***	5.456***	1.081**	2.947**
other tenant	-0.305***	0.737***	-0.609***	0.543***	-3.202***	0.040***
free	-0.018*	0.714*	-0.026	0.973	-0.452	0.635
Dwelling age						
before 1974	base	base	base	base	base	base
1974–1989	0.361***	1.434***	-1.66***	0.190***	1.213**	3.363**
1990–1998	0.278***	1.320***	-0.495***	0.609***	0.625	1.869
after 1998	-0.018	0.981	-1.657***	0.190***	-0.645	0.524
Urban density						
rural	base	base	base	base	base	base
urban	0.064	1.067	0.371**	1.449**	0.216	1.241
Paris area	-0.591***	0.553***	0.372*	1.451*	1.61	5.036
Pseudo $\bar{\rho}^2$	0.3352					
Log-likelihood	-10704					
Observations	14861					
Base category: Direct Electric Heating						

Relative Risk Ratio of one preference over other

* Coefficient significant at 90 confidence level

** Coefficient significant at 95 confidence level

*** Coefficient significant at 99 confidence level

Table 5: Results of MNL for choice of heating system (2013)

Choice alternatives (Frequency of alternatives)	Individual boiler (45%)		Collective boiler (11%)		Heat pump (7%)		Wood (7.5%)	
	Coefficient	RRR [#]	Coefficient	RRR [#]	Coefficient	RRR [#]	Coefficient	RRR [#]
Constant	-2.5***	NA	-6.82***	NA	-2.53*	NA	-0.34	NA
Size of dwelling	0.01***	1.011***	0	1.002	0.02***	1.015***	0.01***	1.011***
Household size	0.15*	1.162*	0.3**	1.346**	0.34***	1.397***	0.31***	1.365***
Age of RP	0.02***	1.016***	0.02***	1.018***	0	0.997	-0.01	0.991
No access to gas grid	-0.41	0.663	0.62	1.851	0.18	1.196	0.36	1.431
No double-glazing	-0.09	0.918	-0.51	0.599	0.27	1.131	-0.09	0.917
No roof insulation	-0.13	0.879	0.12	1.124	0.13	1.142	-0.18	0.839
Climate zone								
Mediterranean	base	base	base	base	base	base	base	base
middle oceanic	-0.26	0.774	0.11	1.117	-0.83**	0.435**	-0.36	0.697
oceanic	0.11	1.118	1.03**	2.788**	-1.1**	0.333**	-0.75	0.470
fresh oceanic	0.8***	2.233***	1.72***	5.589***	-1.46***	0.233***	-0.27	0.766
modified oceanic	0.66***	1.933***	1.27***	3.564***	-1.24***	0.289***	-0.72*	0.485*
semi-continental	1.03***	2.807***	1.97***	7.188***	-0.09	0.914	0.1	1.105
mountain	0.65**	1.922**	1.42***	4.149***	-0.74*	0.479*	-0.58	0.559
Income per month (euro)								
< 1200	base	base	base	base	base	base	base	base
1201–1900	0.23	1.261	0.65*	1.913*	-0.08	0.919	-0.28	0.754
1901–2700	0.22	1.248	0.26	1.298	-0.37	0.688	-0.42	0.657
2701–3800	0.16	1.174	-0.06	0.945	-0.46	0.628	-1.41***	0.245***
> 3800	0.18	1.201	-0.93	0.395	-0.63	0.532	-0.95*	0.386*
Dwelling occupancy status								
free	base	base	base	base	base	base	base	base
owner	0.73	2.081	0.27	1.304	-0.03	0.970	-0.09	0.912
tenant of HLM	1.39***	4.009**	0.83	2.283	-1.24	0.288	-0.67	0.510
other tenant	0.28	1.323	-1	0.367	-1.04	0.354	-1.01	0.362
Dwelling age								
before 1974	base	base	base	base	base	base	base	base
1974–1989	-1.25	0.286	-1.23***	0.291***	-0.75**	0.470**	-0.71**	0.493**
1990–1998	-1.45***	0.234***	-2.68***	0.068***	-0.88*	0.416*	-1.17**	0.309**
after 1998	-1.14***	0.319***	-2.23***	0.098***	0.38	1.465	-0.42	0.655
Dwelling type								
individual house	base	base	base	base	base	base	base	base
apartment	-0.1	0.906	3.95***	51.8***	-0.16	0.852	-1.04***	0.352***
Urban density								
rural	base	base	base	base	base	base	base	base
urban	0.39	1.473	0.77	2.167	0.16	1.177	-0.47	0.627
Paris area	0.01	1.006	1.3*	3.675*	0.33	1.386	-0.79	0.451
Pseudo \bar{r}^2	0.20							
Log-likelihood	-1441.5							
Observations	1350							
Base category: Direct Electric Heating								

Relative Risk Ratio of one preference over other

* Coefficient significant at 90 confidence level

** Coefficient significant at 95 confidence level

*** Coefficient significant at 99 confidence level

7. Who prefers what? Results of post-estimation clustering

In this section, we propose to analyze the barriers to choosing a particular RSHS. More precisely, we attempt to determine the household and dwelling characteristics for which the considered choice alternative has the highest probability of being chosen. The econometric model for choice of RSHS gives us the estimated probability by year and household. The distribution of probability by choice for 2006 is given in Figure 3 (individual houses and apartments) and for 2013 in Figure 4 (all dwellings). For each choice alternative we try to identify the group of households with the highest probability (orange area in the Figures 3–4).

Figure 3: Distribution of estimated probability for individual houses (a) and apartments (b) in 2006

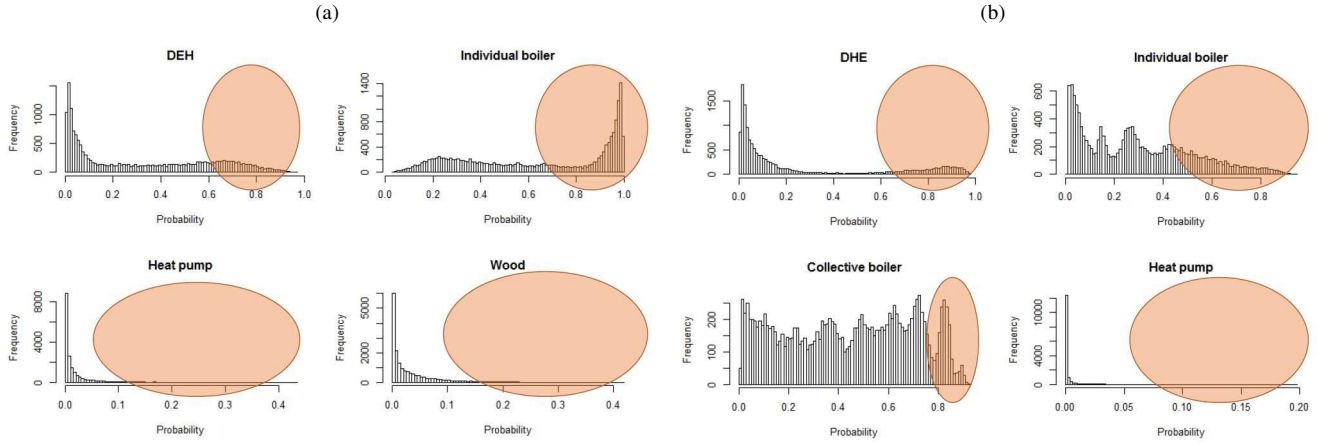
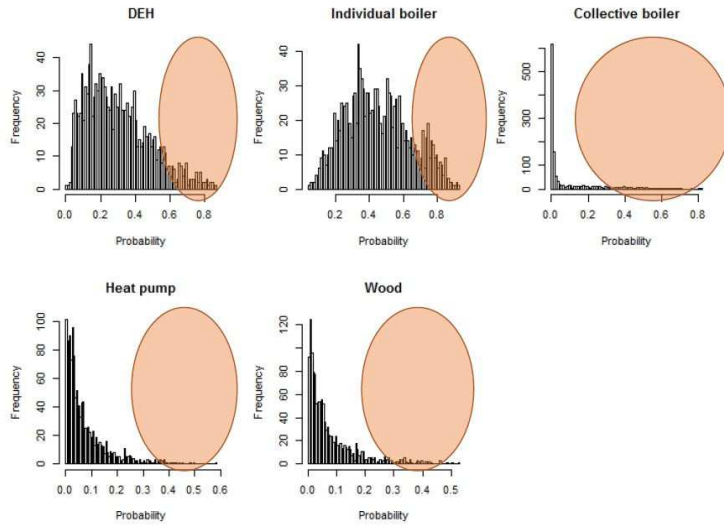


Figure 4: Distribution of estimated probability for dwellings in 2013



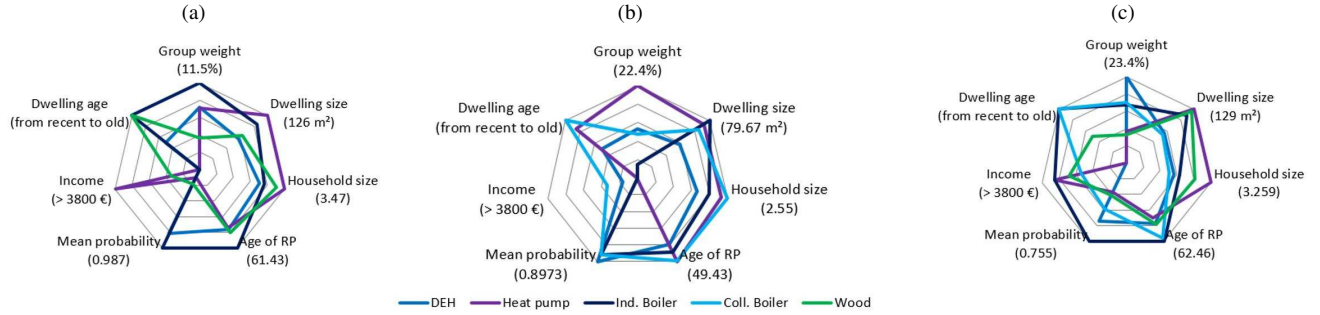
We observe that the probability of using a heat pump or wood heating in 2006 is very low. The probability of using a heat pump is a maximum of 0.48 and 0.2 in houses and apartments respectively. For many dwellings the probability is close to 0. The same observation applies to wood heating. In contrast, the probability of using a boiler or DEH comes close to 1 for some dwellings. This means that dwelling and household characteristics strongly determine the RSHS for these households. In the case of apartments, we clearly observe several distributions for boilers: individual boilers seem to divide into 5 groups of household, while in the case of collective boilers we observe 9 groups, 7–8 of which seem to comprise the same number of households.

Looking at 2013, we observe a significant change in probability distribution. The probability of having a heat pump or wood RSHS comes close to 0.6. For many dwellings, the probability of having a collective boiler is close to 0. In the case of DEH and individual boilers, the probabilities are still high, but lower than in 2006. We also observe the small number of households whose probabilities are close to 0 in comparison with 2006.

7.1. Global results

The results of clustering are given in radar charts Figure 5 showing the mean values of characteristics for each group of household in 2006 and 2013. For example, the mean dwelling size for the heat pump group in 2013 is 129 m² which is double the mean dwelling size for the collective boiler group.

Figure 5: Cluster results for individual houses in 2006 (a), for apartments in 2006 (b) and for all dwellings in 2013 (c)



The detailed results of clustering are given in Appendix A Table A.6. In 2006, the mean probability of having a boiler or DEH by focus group is high compared to the mean probability of the heat pump and wood group: 0.828 for collective boilers in apartments and 0.987 for individual houses. If we consider the number of households with strong constraints, at 0.8 mean probability by cluster, we can conclude that 21% of households living in houses and 36% living in apartments had no choice in 2006²⁰. The situation had improved by 2013: only the individual boiler group has a strong constraint, with a mean probability equal to 0.75. If we look at the dwelling occupation status, we note that the group for heat pumps, individual boilers and wood heating is composed of owners. In contrast, the highest probabilities of having a collective boiler concern social tenants in HLMs and other tenants for DEH. The results are the same for both years.

The mean dwelling size in the DEH focused group is the lowest compared to other groups: 85 m² for the DEH group against 95 m² (wood), 127 m² (individual boiler) and 150 m² (heat pump). In 2013 the mean dwelling size for DEH and collective boiler groups are close to 70 m², whereas the mean dwelling size for other RSHSs is around 120 m². The heat pump group has the largest household size: 3.47 people in 2006 (houses) and 3.259 in 2013, followed by individual boilers and wood heating. In 2006, the collective boiler group has the highest mean household size (2.55) in apartments, whereas in 2013 it has the lowest (1.65). The age of reference person for heat pumps, DEH and wood is similar in all models. High income characterizes the heat pump group in 2006 (houses) and 2013. In 2013, households in the individual boiler group also have the high level of income. Households in the boiler groups (both individual and collective) live in dwellings built before the first thermal regulation (1974).

7.2. Direct electric heating

The group weight for DEH is 8.3% in 2006 (houses) and 23.4% in 2013. The mean probability is very high in groups for 2006: 0.809 for houses and 0.897 for apartments. It is clear that for 20.3% of households, dwelling and household characteristics strongly determine the presence of DEH. The mean probability of opting for another RSHS is between 0 and 0.15. These barriers have been overcome in 2013. The mean probability of having DEH is 0.56, but the group weight is 23.4%. In 2006, DEH is widespread in the Mediterranean and semi-continental climates, whereas in 2013 it also concerns a significant number of households in oceanic and middle oceanic climate zones. The mean size of dwelling and the age of reference person changed little between 2006 and 2013. However, the mean size of household decreased from 2.45 (houses) and 1.82 (apartments) people per household in 2006 to 1.82 for all dwellings in 2013. This is in line with general statistics: household size on average dropped from 2.3 in 2006 to 2.16 in 2013. Household groups are not characterized by income (except apartments in 2006) or urban density (except apartments). Access to the gas grid, age of dwelling, income and urban density do not determine which households prefer direct electric heating in 2013.

²⁰We consider the total weight for all clusters in which the mean probability is higher than 0.8.

7.3. Individual boiler

This groups weight increases from 3.4% in 2006 to 15.9% in 2013, with households residing in all climate zones except for the Mediterranean and middle oceanic. We observe a decrease in household size from 2.65 and 2.17 in 2006 to 2.05 in 2013. As for DEH, this is probably due to the general decrease in household size between the two years. The 2013 group is characterized by high income (2701-3800 €/ month). The mean probability of having an individual boiler is 0.987 (houses) and 0.823 (apartments) in 2006, whereas the mean probability of using another RSHS is close to 0. However, in 2013 the mean probability decreases to 0.755 in favor of DEH with a 0.138 mean probability in these group.

7.4. Collective boiler

Collective boilers were preferred by people with average age of 49 in 2006, but the mean reference person's age was 10 years older in 2013. As in the case of individual boilers, household size decreased between the two years. We do not observe any other changes, except the group weight and mean probability, which was 0.828 in 2006 and rose to 0.446 in 2013, competing with individual boilers, with a mean probability of 0.375 in the same group. Typical households from this group are social tenants on incomes below 1,900 euro per month living in apartments built before 1974 in urban area zone.

7.5. Heat pump

We do not observe any significant differences between the household groups that prefer heat pumps in 2006 and 2013. In both cases, this group is a large young family with a high income, living in a Mediterranean climate zone in a dwelling built after 1998 with a good level of insulation. However, the group for 2013 is more accurate because the mean probability of choosing a heat pump is about 0.22, whereas it was about 0.095 in 2006. In 2006 the mean probabilities for DEH and individual boilers are 0.533 and 0.346 respectively. Looking at 2013, the mean probabilities for other RSHSs are close to that of heat pumps, except for collective boilers (0.0009). We conclude that in 2013 households in this group have no preference between DEH, heat pumps, individual boilers and wood heating.

7.6. Wood heating

Wood heating systems are the preferred choice in the Mediterranean, middle oceanic and oceanic climate zones. The mean dwelling size increased from 95 m² in 2006 to 125 m² in 2013, but household size decreased. In 2006, only low-income households preferred wood. The 2013 group is also characterized by households living in dwellings built before 1974 and after 1998. This may be explained by the spread of wood heating systems from 1% in 2006 to 11.3% in 2013 especially in new buildings. The mean probability of using wood heating is 0.312, followed by individual boilers (mean probability about 0.28) and heat pumps (mean probability about 0.155). The 2013 wood heating group differs from all of the other groups regarding the sum of mean probability, which is 0.783. For the other groups, the sum of probabilities of using all considered RSHSs is close to 1. We can conclude that the households in this group also prefer other RSHSs not mentioned in the study.

8. Conclusion

The aim of this research was to analyze how strong the household situation impacts the likelihood of using different types of RSHS in French dwellings. The quantitative results obtained using econometric methods are consistent with the qualitative knowledge of experts from the energy sector in France. Comparing the results from both years used in the study (2006 and 2013), we identified limited differences in the models. However, we observed that the order of preferences between choice alternatives changed from 2006 to 2013: the probability of choosing wood or a heat pump as a main heating system significantly increased in 2013.

Focusing our study in socio-demographic, dwelling and spatial determinants, we show that in 2006 these characteristics strongly determine the type of RSHS for 21% and 36% of households in, respectively, individual houses and apartments. Thus, these households did not seem to have a choice. The situation had improved by 2013, especially for the wood heating focused group, for whom wood heating is the first choice. However this progression is not sufficient, because individual boilers are the second choice for households in all focused groups.

Climate zone seems to be an important determinant of household choice. This suggests that different government incentive policies could be adopted for each climate zone. EM clustering also showed that most focused groups comprise owners, except for collective boilers and DEH in 2013.

We intend to use these results in a model that describes the home renovation market, i.e. insulation and replacement of RSHSs. This type of model could also be used to calibrate energy models or calculate expected energy consumption (Braun, 2010; Nesbakken, 1999). We use the revealed preferences data, which is related to actual choices in real-life situations, but do not give the information on no selected choice alternatives. To improve our study, we will need to conduct the choice experiment on French households. This will give us a stated preferences dataset able to take into account the attributes of all choice alternatives, calculate willingness-to-pay (Goett et al., 2000), and consider the taste heterogeneity of consumers behavior (Islam and Meade, 2013). These two types of data can be combined to make the choice experiment more realistic (Ben-Akiva and Morikawa, 2002).

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Appendix A. Results of EM clustering

Table A.6: Results of EM clustering

RSHS	Database	Class characteristics													Mean of estimated probability by group and by RSHS				
		Class size	Climate zone ¹	Mean dwelling size	Mean household size	Age of Reference Person	Access to gas grid	Income per month	Dwelling occupancy status	Dwelling age	Dwelling type	Urban density	Double glazing	Roof insulation	DEH	Individual boiler	Collective boiler	Heat pump	Wood
DEH	House 2006 Apt. 2006 2013	8.3%	1,5	85 m ²	2.45	47	no	Owner/Tenant	Tenant	1974–1989	–	–	yes	yes	0.809	0.151	–	0.013	0.024
		12%	1,5	47 m ²	1.82	39	no	< 1200	Tenant	After 1998	–	Urban	yes	–	0.897	0.042	0.06	0	–
		23.4%	1,2,3	73 m ²	1.82	49	–	–	Tenant	–	–	–	yes	–	0.56	0.275	0.04	0.06	0.056
Ind. boiler	House 2006 Apt. 2006 2013	11.5%	3,4,5	127 m ²	2.65	61	yes	–	Owner	Before 1974	–	Urban	–	yes	0.01	0.987	–	0	0.002
		3.4%	2,3,6	80 m ²	2.17	44	yes	–	–	After 1998	–	Urban	yes	–	0.067	0.823	0.109	0	–
		15.9%	4,5,6	116 m ²	2.05	62	yes	2701–3800	Owner	Before 1974	House	–	yes	yes	0.138	0.755	0.017	0.04	0.05
Col. boiler	Apt. 2006 2013	10.7%	3,4,5,6,7	67 m ²	2.74	49	yes	<1900	HLM	Before 1974	–	Paris	yes	–	0.025	0.144	0.828	0.001	–
		16.4%	3,4,5,6,7	69 m ²	1.65	60	yes	1201–1900	HLM	Before 1974	Apartment	Urban	yes	no	0.015	0.375	0.446	0.013	0.011
Heat Pump	House 2006 Apt. 2006 2013	8.2%	1,2	150 m ²	3.48	46	no	> 2701	Owner	After 1998	–	Urban	yes	yes	0.533	0.346	–	0.095	0.02
		22.4%	1,5	74 m ²	2.55	49	–	–	Owner/HLM	Before 1989	–	Paris	–	–	0.319	0.234	0.437	0.008	–
		8.6%	1	130 m ²	3.26	44	–	2701–3800	Owner	After 1998	House	Urban	yes	yes	0.247	0.3	0.009	0.28	0.163
Wood	House 2006 2013	4.2%	1,2,3	95 m ²	3.14	50	no	<1900	Owner	Before 1974	–	Rural	–	–	0.364	0.452	–	0.006	0.176
		7.7%	1	125 m ²	2.65	49	no	< 2700	Owner	<1974 or >1998	House	Rural	yes	yes	0.024	0.287	0.005	0.155	0.312

¹ Mediterranean (1), middle oceanic (2), oceanic (3), fresh oceanic (4), modified oceanic (5), semi-continental (6), mountain (7).